







# A new variational family for Bayesian phylogenetics

Lloyd T. Elliott<sup>1</sup>, Evan Sidrow<sup>1</sup>, Alexandre Bouchard-Côté<sup>2</sup>

<sup>1</sup>Simon Fraser University, <sup>2</sup>University of British Columbia

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#### **Genetics**

- COVID-19 RNA genome:
  - AUUAAAGGUUUAUACCUUCC ...

SNP: Single Nucleotide Polymorphism (basepair substitution)

- Human DNA genome:
  - TAACCCTAACCCTAACCCTA ...

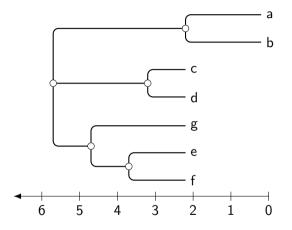
At what time did two observed genetic sequences coalesce?

(We observe N genetic sequences)

#### Related problems:

- Infer mutation rates
- Discover variants of interest/variants of concern (clade emergence)
- Impute missing or ancestral sequences

# **Phylogenetics**



- Leaf nodes: Observed RNA sequences
- Interior nodes: Unobserved
- Goal: Infer tree topology and branch lengths

## **Bayesian phylogenetics**

- Place a prior on trees (Kingman's coalescent), develop proposals (generalized stepping-stone sampling), perform MCMC (BEAST)
  - Largest dataset studied with BEAST: ~25k taxa (L. Lyu et al. PNAS 2025)

- Construct a variational family, perform variational Bayes (vB: more scalable?)
  - vB is an iterative inference algorithm that approximates the posterior using a family of functions (in contrast, MCMC approximates with a set of samples)

 Note: These methods both require a likelihood function to link the tree topology and branch lengths to the observed sequences (JC, K2P, GTR)

# Single-linkage clustering

Construct a rooted tree topology from a distance matrix:

- 1: **Input:** Distances  $T \in \mathbb{R}_{>0}^{\binom{N}{2}}$  and taxa set  $\mathcal{X} = \{\{x_1\}, \{x_2\}, \dots, \{x_N\}\}.$
- 2: **for** n = 1, ..., N 1 **do**
- 3:  $w^*, z^* \leftarrow \arg\min_{w,z} \{t^{\{w,z\}} : w,z \text{ have not coalesced by the } (n-1)\text{-st event}\}$
- 4: Set  $W_n \in \mathcal{X}$  to be the set containing  $w^*$
- 5: Set  $Z_n \in \mathcal{X}$  to be the set containing  $z^*$
- 6:  $t_n \leftarrow t^{\{w^*,z^*\}}$
- 7: Remove  $W_n$ ,  $Z_n$  from  $\mathcal{X}$  and add  $W_n \cup Z_n$  to  $\mathcal{X}$
- 8: end for
- 9:  $\tau \leftarrow \{\{W_n, Z_n\}\}_{n=1}^{N-1}$
- 10:  $t \leftarrow \{t_n\}_{n=1}^{N-1}$
- 11: Return  $(\tau, t)$

#### **Notation**

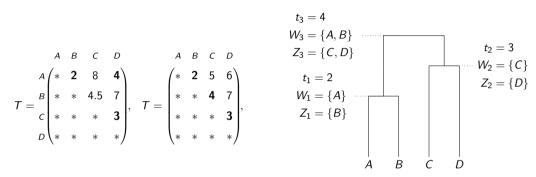
- $\{W_n, Z_n\}$  is a bipartition (a.k.a. subsplit)
  - This means the *n*-th coalescent event involves a clade with leaves  $W_n$ , and a clade with leaves  $Z_n$ .  $(W_n, Z_n \subseteq \{x_1, \dots x_N\}, W_n \cap Z_n = \emptyset)$

- $\{\{W_n, Z_n\}\}_{n=1}^{N-1}$  is a sequence of bipartitions
  - $\circ$  The topology of a tree with N leaves can be uniquely described by N-1 bipartitions
  - $\circ \#W_1 = \#Z_1 = 1, W_{N-1} \bigcup Z_{N-1} = \{x_1, \ldots, x_n\}$

• For  $n \leq N-1$ , w and z have coalesced by the n-th event if there is a bipartition  $W_m, Z_m$  with  $w \in W_m, z \in Z_m$  and  $m \leq n$ 

# Single-linkage clustering (cont)

$$T = \begin{pmatrix} A & B & C & D \\ A & * & 2 & 8 & 4 \\ * & * & 4.5 & 7 \\ C & * & * & * & 3 \\ D & * & * & * & * \end{pmatrix}, \quad T = \begin{pmatrix} * & 2 & 5 & 6 \\ * & * & 4 & 7 \\ * & * & * & 3 \\ * & * & * & * \end{pmatrix},$$



Example: Two matrices T result in the same phylogenetic tree after running single-linkage clustering. Entries of T that trigger coalescence are bolded

## Related work: GeoPhy

- Associate *i*-th taxa with a point  $x_i$  in k-dimensional space
- The variational family is a distribution q on  $X \in \mathbb{R}^{N \times D}$
- Given a draw X, the corresponding tree is found by forming the distance matrix T such that  $t^{(i, j)}$  is the Euclidean distance between  $x_i$  and  $x_j$ , and then running single-linkage clustering
- It is difficult to find a closed form for the distribution on tree topologies induced by q, so the variational family is represented by samples
- The method is generalized to points on a hyper-sphere, with distances given by geodesics

— T. Mimori, M. Hamada. NeurIPS 2023

#### Related work: VBPI

- ... Variational Bayesian Phylogenetic Inference
- Run an initializing MCMC in BEAST
- Record all subsplits (bipartitions) appearing in the initializing MCMC run
- Form subsplit Bayesian network (SBN) from these subsplits
- The variational family is a distribution on the SBNs
- Excellent accuracy in likelihood estimation
- May not be scalable, as number of subsplits in the MCMC run is large

— C. Zhang, F.A. Matsen IV. ICLR 2019

# **VIPR: Variational inference with products** ...

... over bipartitions

- Like GeoPhy, we consider single-linkage clustering to map from distance matrices to a tree
- We place the variational family as a distribution directly on the distance matrix (each off diagonal entry of the upper triangle is log normal)
- We derive the distribution of a tree implied by the variational family
- This distribution has a closed form as a sum over bipartitions (subsplits)

— E. Sidrow, A. Bouchard-Côté and L.T. Elliott. ICML 2025

#### The probability of a tree has a closed form ...

 $\dots$  in terms of the distribution of the distance matrix T

**Proposition.** If the random variables  $t^{\{u,v\}}$  are mutually independent, and all  $q_{\phi}^{\{u,v\}}$  are continuous in  $\phi$  and t for all  $\{u,v\}$  with  $u,v\in\mathcal{X}$ , and  $Q_{\phi}^{\{u,v\}}$  is the survival function of  $t^{\{u,v\}}$ , then  $q_{\phi}(\tau,t)$  has the following form:

$$q_{\phi}(\tau,t) = \prod_{n=1}^{N-1} \left( \left( \sum_{\substack{w \in W_n \\ z \in Z_n}} \frac{q_{\phi}^{\{w,z\}}(t_n)}{Q_{\phi}^{\{w,z\}}(t_n)} \right) \prod_{\substack{w \in W_n \\ z \in Z_n}} Q_{\phi}^{\{w,z\}}(t_n) \right).$$

Here  $W_n, Z_n$  is the bipartition induced by the *n*-th coalescent,  $\tau$  is the tree topology, and t are the coalescent times

#### Intuition

- If two clades coalesce at time  $t_n$ , one pair (with one taxa from one clade, the other taxa from the other clade) must coalesce at time  $t_n$  in the distance matrix. And all other such pairs must coalesce after time  $t_n$  in the distance matrix (survival function  $Q(\cdot) = \Pr(rv \ge \cdot)$ ). This marginalizes which pair coalesced
- Since entries in the distance matrix are independent, the above marginalization is summed over coalescent events

#### Complexity

$$\prod_{n=1}^{N-1} \left( \left( \sum_{\substack{w \in W_n \\ z \in Z_n}} \frac{q_{\phi}^{\{w,z\}}(t_n)}{Q_{\phi}^{\{w,z\}}(t_n)} \right) \prod_{\substack{w \in W_n \\ z \in Z_n}} Q_{\phi}^{\{w,z\}}(t_n) \right).$$

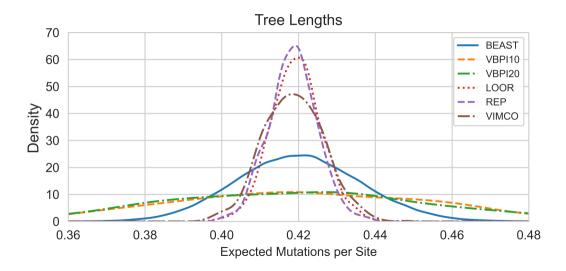
- Each pair w,z occurs once in one of the inner sums, and once in one of the inner products:  $\mathcal{O}(N^2)$  operations. There are N-1 terms in total, so an additional  $\mathcal{O}(N)$  operations
- Total complexity for evaluating q or derivatives of q:  $\mathcal{O}(N^2 + N) = \mathcal{O}(N)$

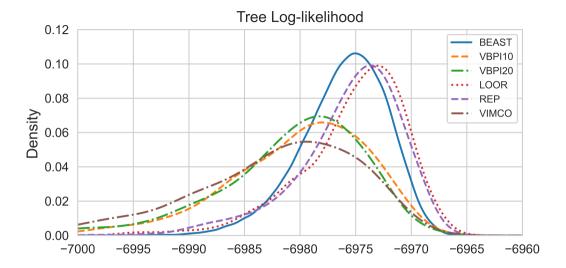
#### Inference

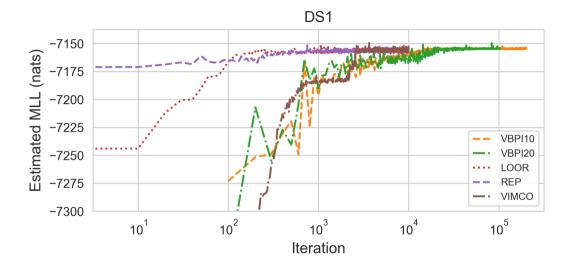
 Derivatives of the sum product in the proposition are found using automatic differentiation (torch/autodiff)

- ELBO optimization is done using:
  - o VIMCO (Mnih, Rezende 2016)
  - LOOR
  - $\circ\,$  or the reparameterization-trick

# **Results: DS1** (N = 27, M = 1949)

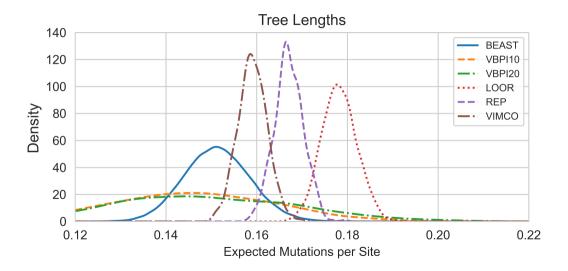


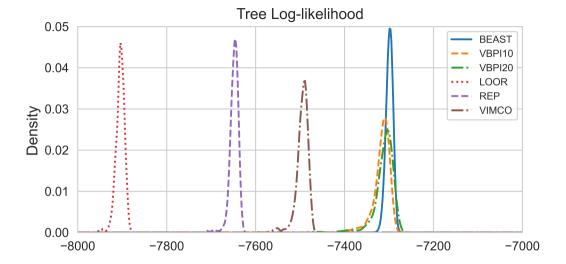




Thank You!

# **Results: COV (**N = 72, M = 3101**)**





#### **COVID-19 Dataset**

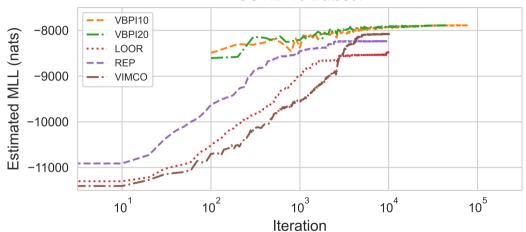


Table 3. Number of tree structure parameters versus number of taxa (NTAXA) on simulated data with 1,000 sites.

NTAXA	VBPI	VIPR
8	4	56
16	44	240
32	55	992
64	3,826	4,032
128	29,939	16,256
256	127,217	65,280
512	319,533	261,632